

# Long CPI Wideband GMTI

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**Abstract** The conventional approach to GMTI uses narrowband signals and a short coherent processing interval (CPI). In this talk, we examine some of the fundamental theoretical issues involved in GMTI with wideband signals and long CPIs (WL-GMTI). The possibility of wideband, long CPI GMTI has received some attention in recent years, and there are a number of potential benefits:

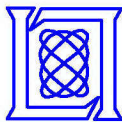
- 1) Improved minimum detectable velocity (MDV).
- 2) Detection of targets with zero radial velocity (but non-zero tangential velocity).
- 3) Better fit with dual-use SAR/GMTI architectures.
- 4) Less demanding array requirements (shorter and/or sparser arrays).
- 5) Greater robustness to clutter internal motion.

The most convenient framework for WL-GMTI is a “post-SAR” architecture, where each spatial channel is pre-processed with synthetic aperture radar (SAR) image processing. The post-SAR architecture is the natural generalization of post-Doppler STAP to the wideband, long-CPI case.

Exact steering vectors in the post-SAR framework are computed analytically for constant-velocity targets, assuming a calibrated array. The steering vectors can be used with algorithms such as the GLRT or AMF to perform adaptive detection on the post-SAR data. We also derive a simple, exact expression for SINR loss when the covariance is known exactly. The loss is a two-dimensional function of both target velocity components, indicating the capability to detect both radial and non-radial target motion.

The final section of this talk examines WL-GMTI performance bounds based on optimal Bayesian detection. In particular, we study how detection performance varies as a function of the number of pixels that the moving target “smears” over in the SAR image. There is a surprising improvement in detection performance when the clutter has strong non-Gaussian tails. In at least some cases, it appears that much of the performance can be achieved with a simple sub-optimal detector.

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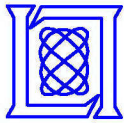
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# Wideband, Long-CPI GMTI

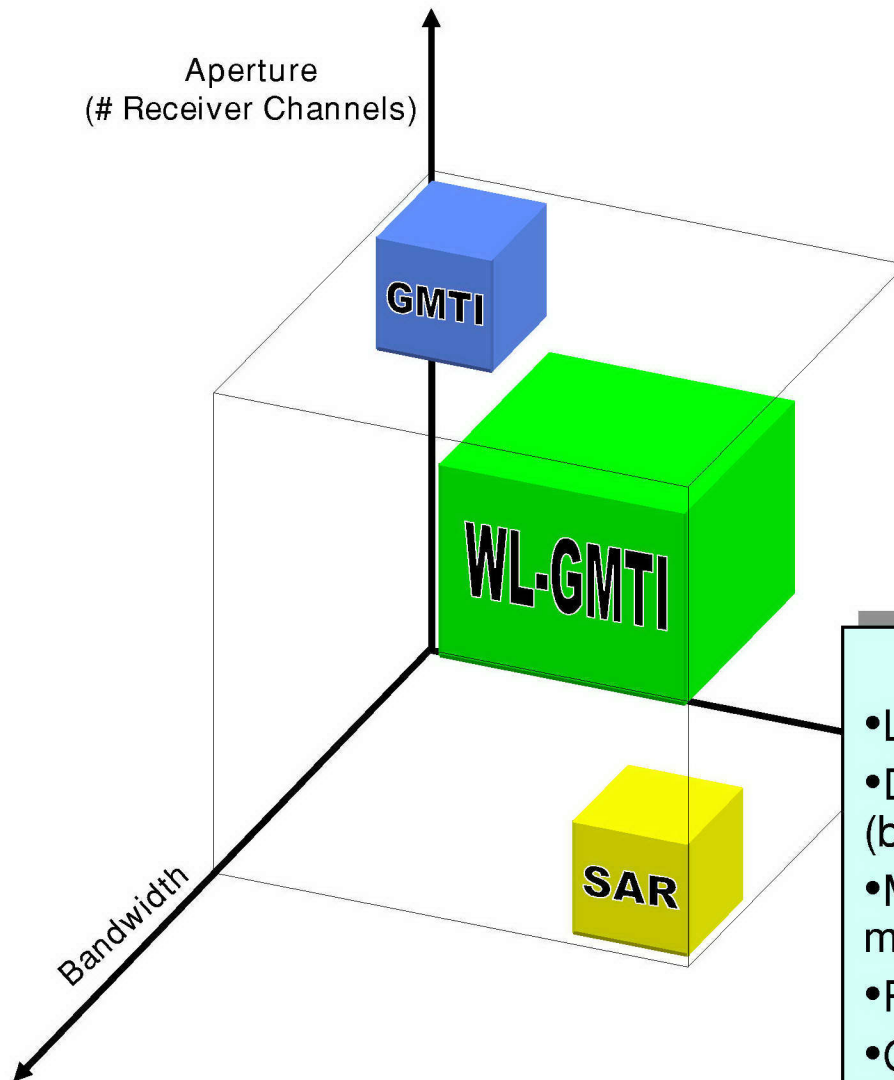
Ali F. Yegulalp

12<sup>th</sup> Annual ASAP Workshop  
16 March 2004

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# Wideband, Long-CPI GMTI (WL-GMTI)



## Premise:

Combine physical aperture of GMTI with bandwidth and integration time of high-resolution SAR.

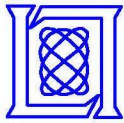
## Purpose:

Supplement traditional SAR and GMTI modes by detecting slow and/or low RCS moving targets.

## Potential Benefits:

- Lower minimum detectable velocity (MDV)
- Detection of targets with zero radial velocity (but non-zero tangential velocity)
- More efficient use of radar resources in multi-mode SAR/GMTI platforms
- Reduced array requirements (sparser/shorter)
- Greater robustness to clutter internal motion

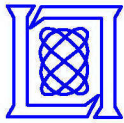




# Background and References

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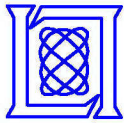
- J.K. Jao, J. Tsay, and S. Ayasli, "Single-Aperture SAR Detection of Moving Targets," *Proc. 2001 MSS Tri-Service Radar Sym.*, John Hopkins Univ., Maryland, May 2001 (SECRET).
- J. Franz, A. Yegulalp, J.K. Jao, and Serpil Ayasli, "Adaptive Airborne Radar Detection of Moving Targets Under Foliage," *Proc. 2001 MSS Tri-Service Radar Sym.*, John Hopkins Univ., Maryland, May 2001 (SECRET).
- J.K. Jao, "Theory of Synthetic Aperture Radar Imaging of a Moving Target," *IEEE Trans. on Geoscience and Remote Sensing*, vol. 39, no. 9, pp. 1984-1992, September 2001.
- J.K. Jao, T.J. Murphy, "Results of the Foliage Penetration Radar and Electronic Support Measures Synergy for Targeting (FOREST) Test Bed Point Design Study," Lincoln Laboratory Project Report FPR-9, October 2001.
- A.F. Yegulalp, "FOPEN GMTI Using Multi-Channel Adaptive SAR," *10<sup>th</sup> Annual ASAP Workshop*, Lincoln Laboratory, Lexington, Massachusetts, March 2002.
- A.F. Yegulalp, "Analysis of SAR Image Formation Equations for Stationary and Moving Targets," Lincoln Laboratory Project Report FPR-14, June 2002 (Distribution Unlimited).
- J.K. Jao, A.F. Yegulalp, J.R. Franz, and S. Ayasli, "New Results of Airborne Multi-Channel Radar Detection of Moving Targets Under Foliage," *48<sup>th</sup> Tri-Service Radar Sym.*, Naval Post-Graduate School, Monterey, California, June 2002 (SECRET).



# Purpose of this Talk

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- Develop a basic framework for discussing and analyzing WL-GMTI
- Show how some of the basic tools of adaptive processing translate to WL-GMTI
  - Steering vectors
  - SINR loss
  - Detection
- Stimulate further interest and work!



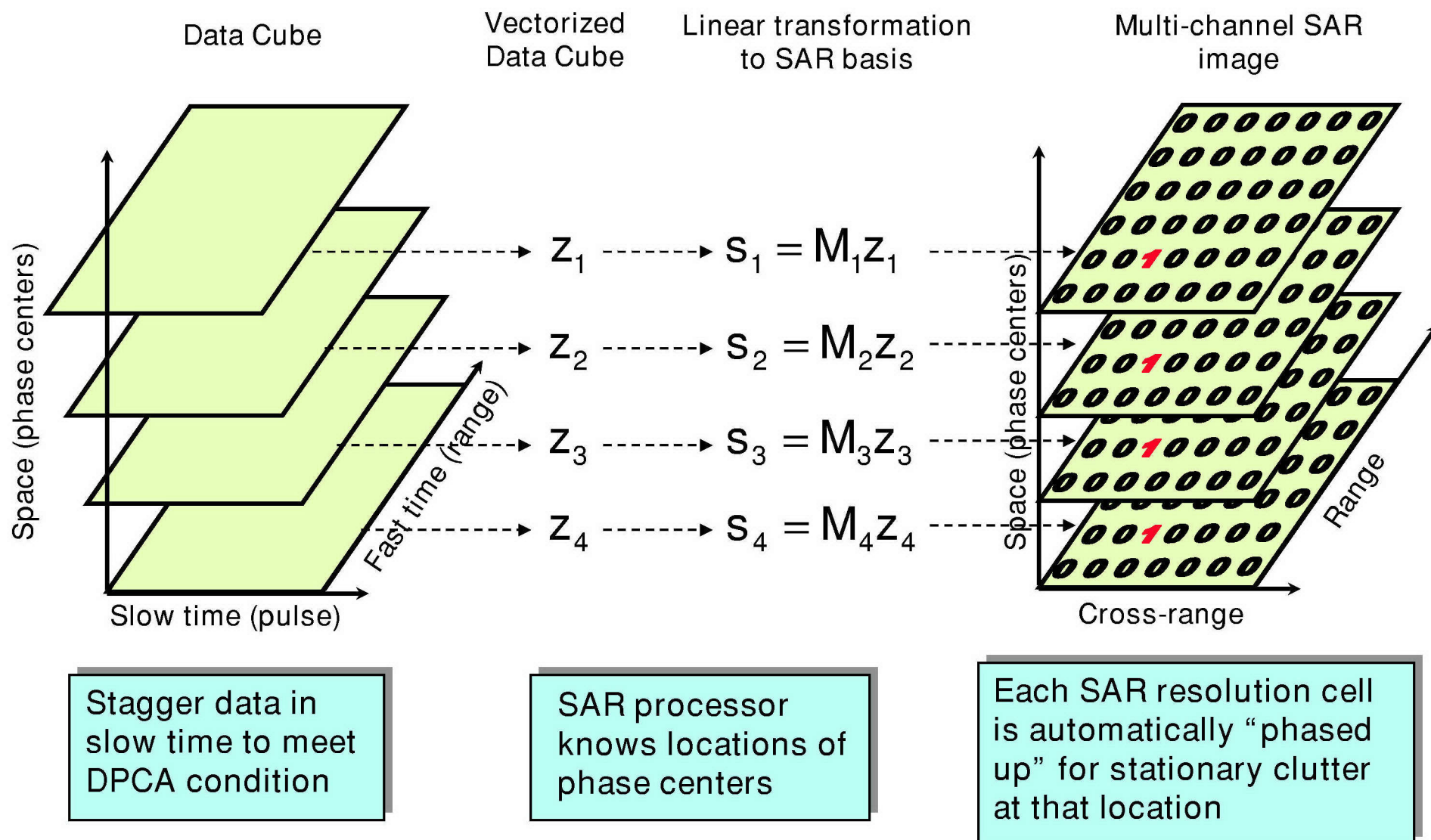
# Outline

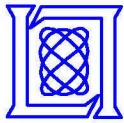
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- SAR as the WL-GMTI pre-processor
- WL-GMTI steering vectors
- SINR loss prediction for WL-GMTI
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# SAR Pre-processing



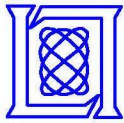


# Properties

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- SAR is the wideband, long-CPI generalization of the Doppler processor in ordinary post-Doppler STAP
- Linear transform of input data cube
  - Annihilates the exoclutter subspace
  - Invertible transformation of the endoclutter subspace
- “Freezes” clutter into SAR resolution bins
  - Clutter in one bin is well-decorrelated from other bins multiple resolution cells away.
  - Stationary targets have trivial steering vectors
  - Stationary clutter has trivial covariance
- Moving targets smear over multiple resolution cells

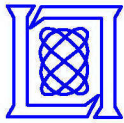




# Benefits of High Resolution for GMTI

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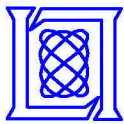
- Target-to-clutter and target-to-noise ratio improve with increasing resolution
  - Improvement holds at least until target is resolved
  - Improvement can continue further if target contains small dominant scatterers
- High spatial resolution provides more clutter per unit area for training adaptive processor
  - Can train in both range and azimuth
- Abundance of training data facilitates more powerful adaptive processing methods
  - Algorithms with more adaptive DOFs
  - Automated data editing to eliminate potential movers from training data



# Outline

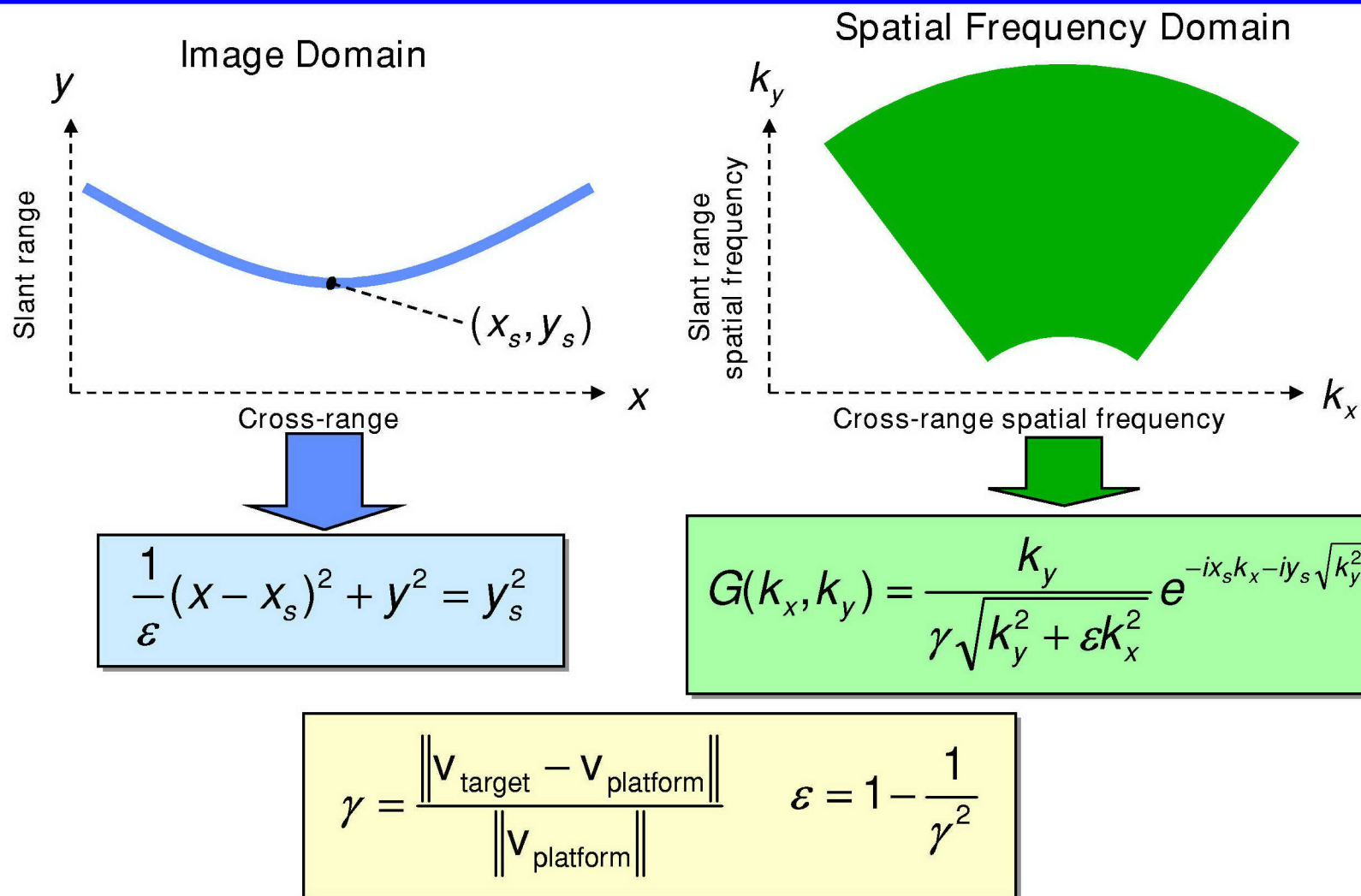
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# SAR Steering Vector\*

## Constant Velocity Point Target



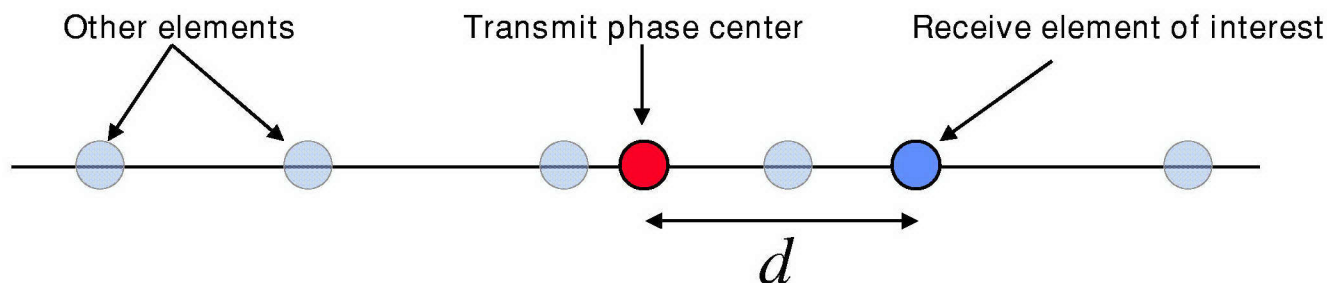
\*"Analysis of SAR Image Formation Equations for Stationary and Moving Targets," A.F. Yegulalp, Lincoln Laboratory Project Report FPR-14, 20 June 2002 (Distribution Unlimited)





# WL-GMTI Steering Vector

## Constant Velocity Point Target



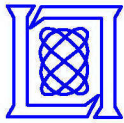
$$x_s(d) = x_s(0) + \frac{d}{2} \left( 1 - \frac{1}{\gamma} \cos \psi \right)$$

$$y_s(d) = y_s(0) - \frac{d}{2} \sin \psi$$

$$\gamma = \frac{\| \mathbf{v}_{\text{target}} - \mathbf{v}_{\text{platform}} \|}{\| \mathbf{v}_{\text{platform}} \|} \quad \varepsilon = 1 - \frac{1}{\gamma^2}$$

$$\cos \psi = \frac{(\mathbf{v}_{\text{platform}} - \mathbf{v}_{\text{target}}) \cdot \mathbf{v}_{\text{platform}}}{\| \mathbf{v}_{\text{platform}} - \mathbf{v}_{\text{target}} \| \| \mathbf{v}_{\text{platform}} \|}$$

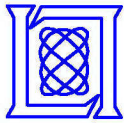
$$G(k_x, k_y; d) = \frac{k_y}{\gamma \sqrt{k_y^2 + \varepsilon k_x^2}} e^{-ix_s(d)k_x - iy_s(d)\sqrt{k_y^2 + \varepsilon k_x^2}}$$



# Outline

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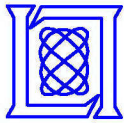
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- **SINR loss prediction for WL-GMTI**
  - Theory
  - Examples
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# Simplifying Assumptions

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- Large clutter-to-noise ratio
- Elements are mutually calibrated
- No internal clutter motion, crab, unmeasured aircraft motion and vibration
- No jammers and other interference
- Isotropic element patterns
- Optimal AMF processing with perfect knowledge of steering vectors and clutter covariance



# Clutter-Limited SINR Loss Calculation

Target steering vector

$$\mathbf{u} = \sum_{n=1}^N \mathbf{v}(n) \otimes \mathbf{e}(n)$$

Vectorized target image from  $n$ -th channel

Element space basis vector

$$\mathbf{e}(1) = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \mathbf{e}(2) = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \dots \mathbf{e}(N) = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$$

Clutter + noise covariance

$$\mathbf{R} = \sigma_n^2 \mathbf{I} \otimes \mathbf{I} + \mathbf{R}_c \otimes \mathbf{e}\mathbf{e}^H$$

Noise power

Clutter pixel-space covariance

$$\mathbf{e} = \sum_{n=1}^N \mathbf{e}(n)$$

Inverse covariance

$$\mathbf{R}^{-1} = \frac{1}{\sigma_n^2} \mathbf{I} \otimes \mathbf{I} - \frac{1}{\sigma_n^4} \left( \mathbf{R}_c \left( \mathbf{I} + \frac{N}{\sigma_n^2} \mathbf{R}_c \right)^{-1} \right) \otimes \mathbf{e}\mathbf{e}^H$$

Large clutter-to-noise limit

$$\sigma_n^2 \mathbf{R}^{-1} \xrightarrow{\sigma_n^2 \rightarrow 0} \mathbf{I} \otimes \mathbf{I} - \frac{1}{N} \mathbf{I} \otimes \mathbf{e}\mathbf{e}^H$$

SINR loss

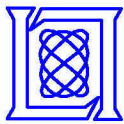
$$\sigma_n^2 \frac{\mathbf{u}^H \mathbf{R}^{-1} \mathbf{u}}{\mathbf{u}^H \mathbf{u}} = 1 - \frac{\left\| \sum_{n=1}^N \mathbf{v}(n) \right\|^2}{N \sum_{n=1}^N \left\| \mathbf{v}(n) \right\|^2}$$



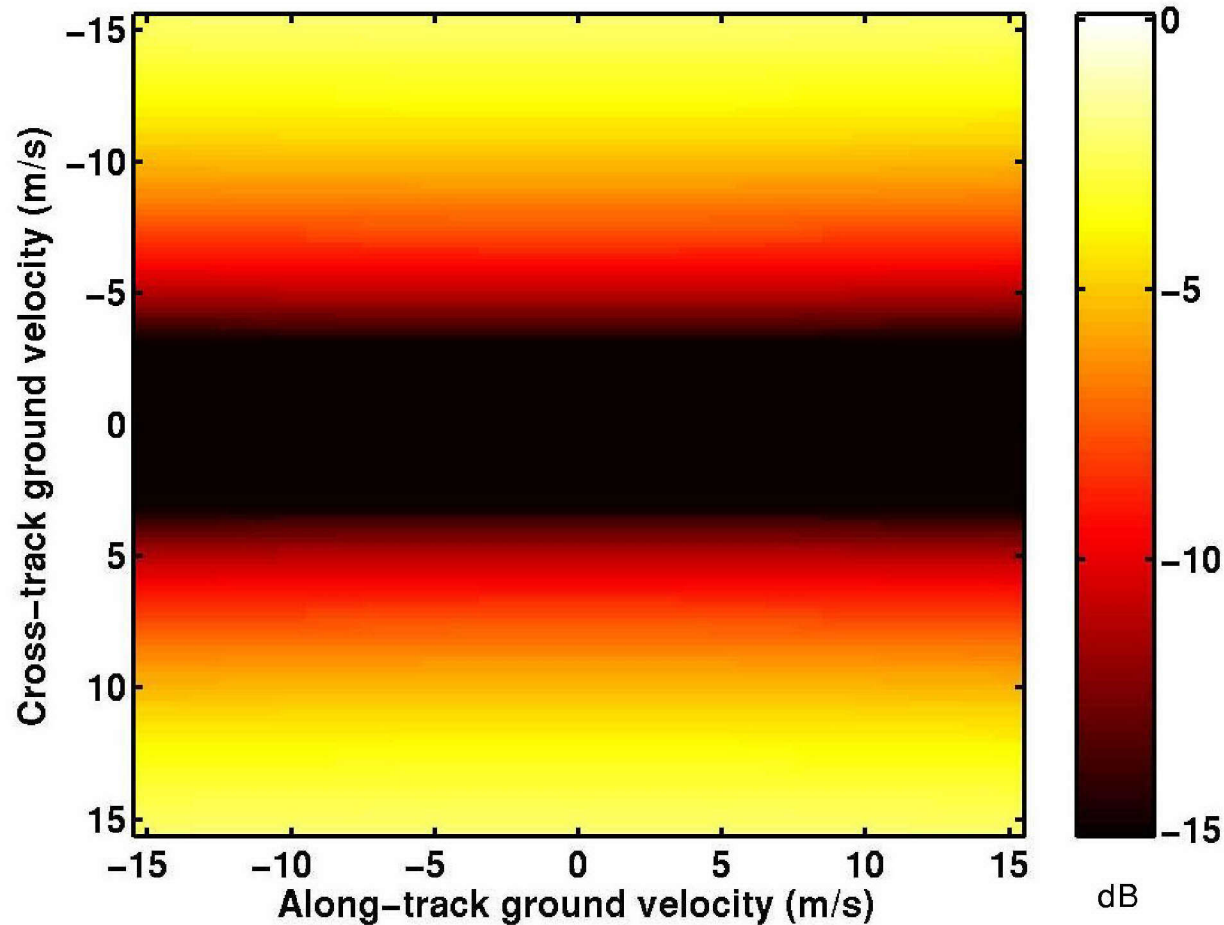
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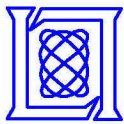
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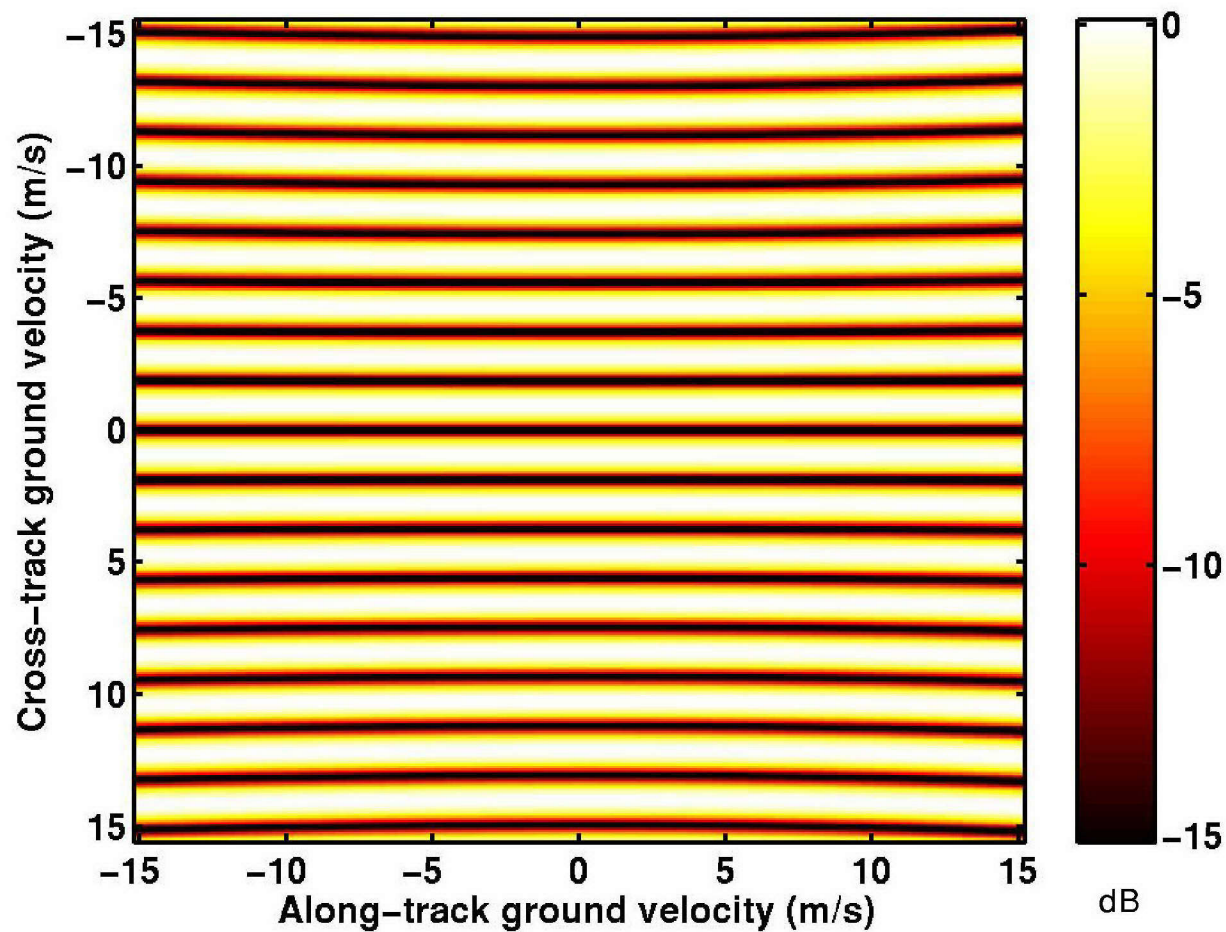
# SINR Loss Example



- Carrier = 1 GHz
- Bandwidth = 10 MHz
- CPI = 50 ms
- Range = 10 km
- Airspeed = 180 m/s
- Phase center locations = [0,1] m

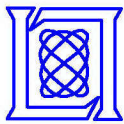


# SINR Loss Example

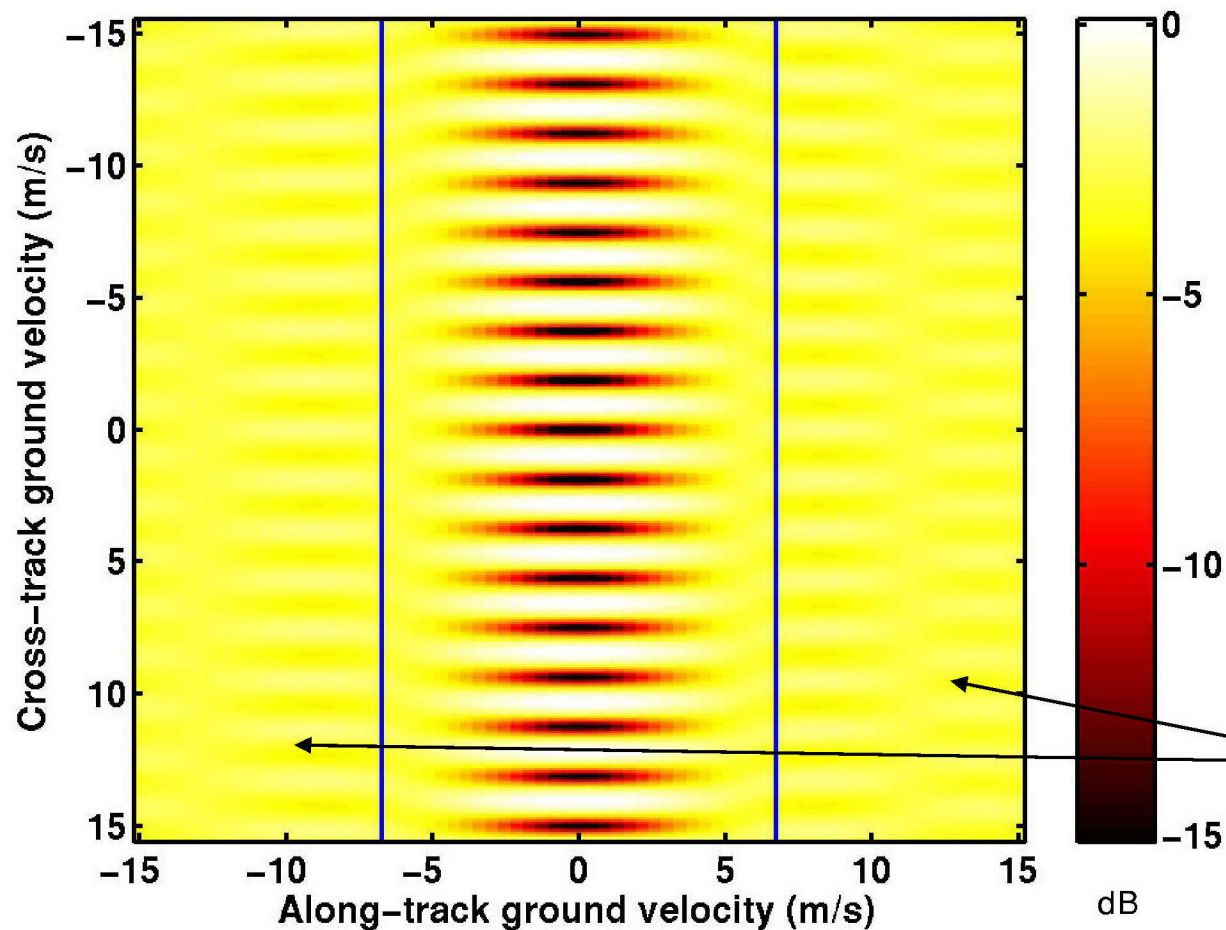


- Carrier = 1 GHz
- Bandwidth = 10 MHz
- CPI = 50 ms
- Range = 10 km
- Airspeed = 180 m/s
- Phase center locations = [0,30] m





# SINR Loss Example

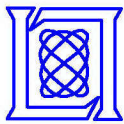


- Carrier = 1 GHz
- Bandwidth = 10 MHz
- CPI = 15 s
- Range = 10 km
- Airspeed = 180 m/s
- Phase center locations = [0,30] m

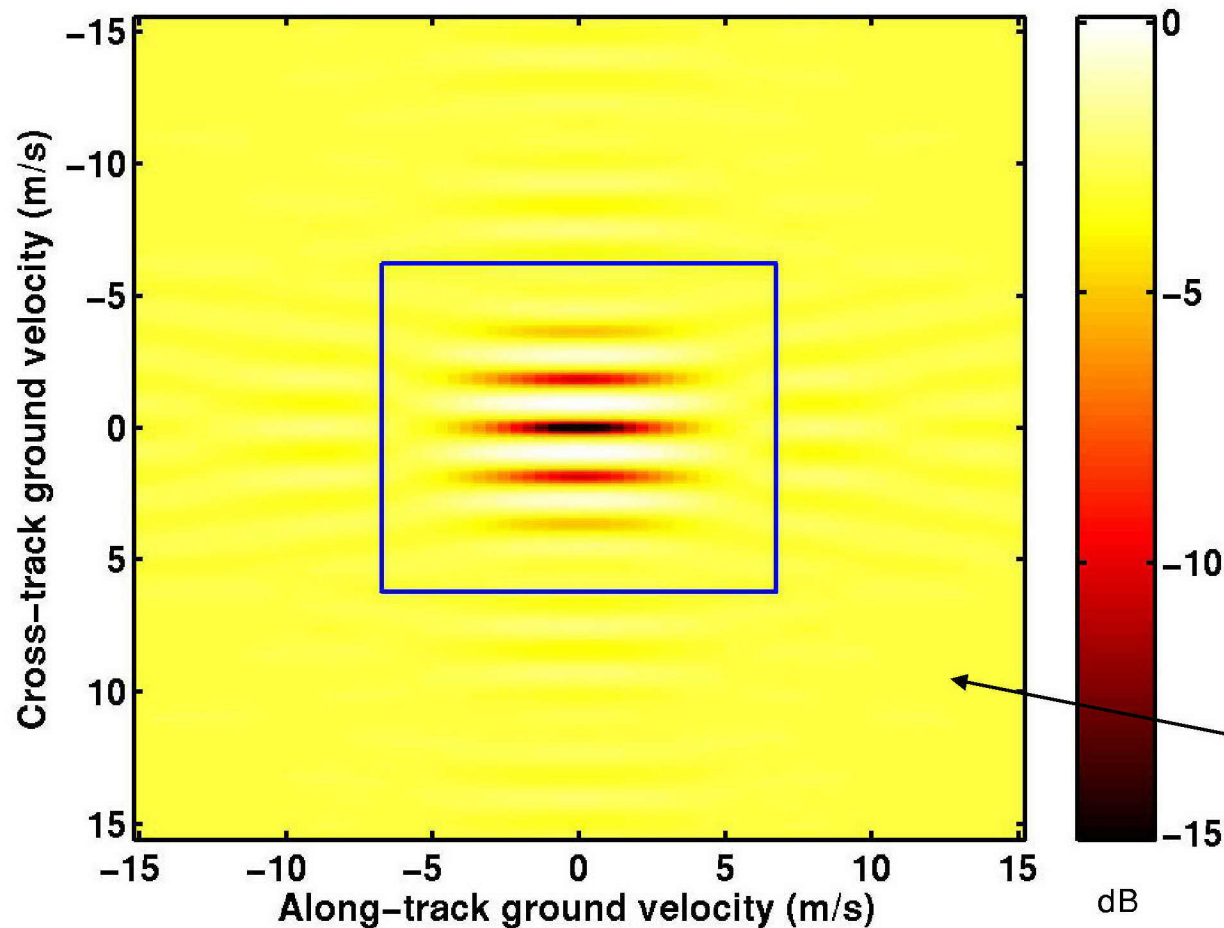
“Differential SAR” zone:

Target moves  $> 2$  SAR resolution cells over time it takes platform to travel the length of the real aperture





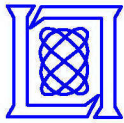
# SINR Loss Example



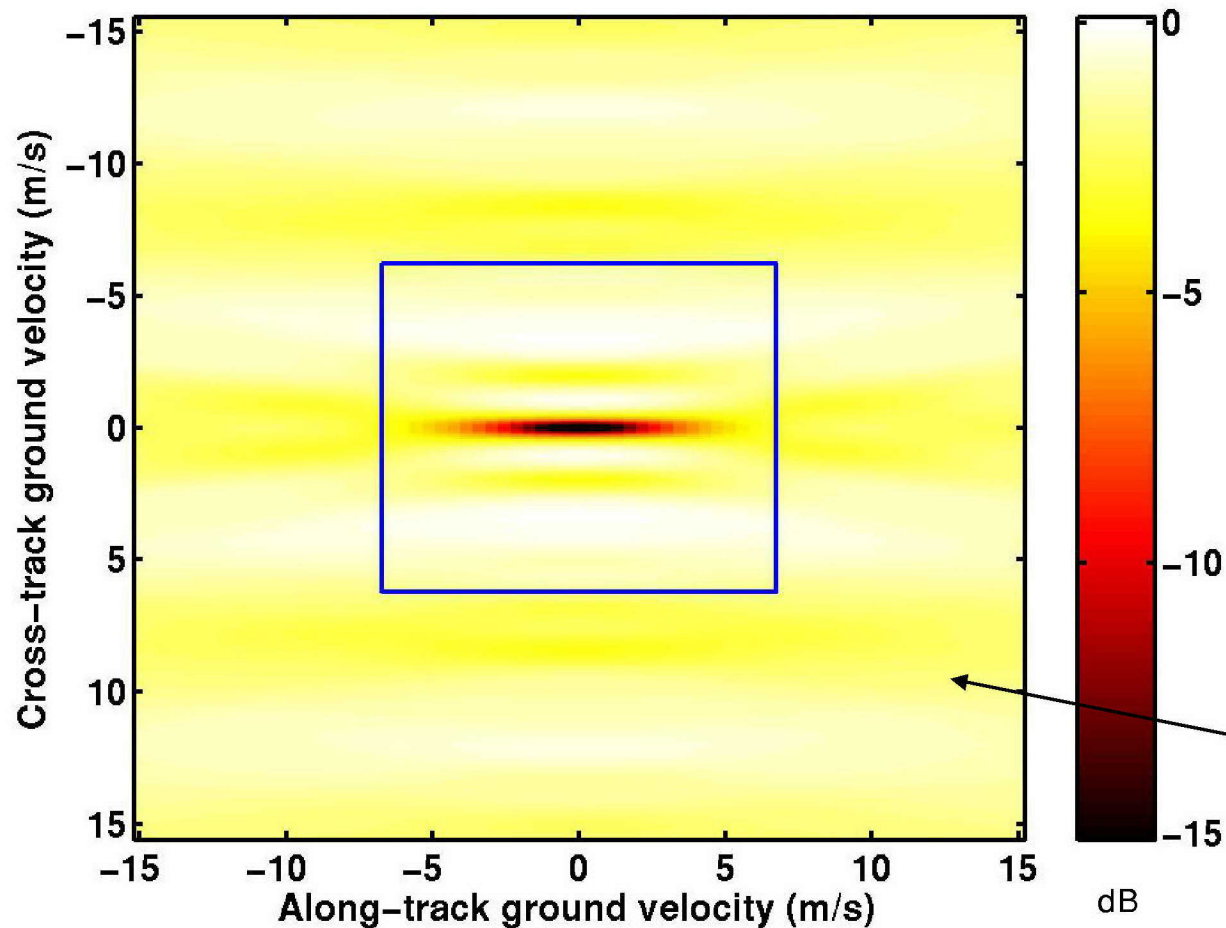
- Carrier = 1 GHz
- Bandwidth = 300 MHz
- CPI = 15 s
- Range = 10 km
- Airspeed = 180 m/s
- Phase center locations = [0,30] m

“Differential SAR” zone:

Target moves  $> 2$  SAR resolution cells over time it takes platform to travel the length of the real aperture



# SINR Loss Example



- Carrier = 1 GHz
- Bandwidth = 300 MHz
- CPI = 15 s
- Range = 10 km
- Airspeed = 180 m/s
- Phase center locations =  $[0, 7, 30]$  m

“Differential SAR” zone:

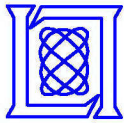
Target moves  $> 2$  SAR resolution cells over time it takes platform to travel the length of the real aperture



# Outline

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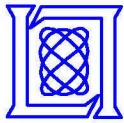
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# SINR Loss vs. Detection Performance

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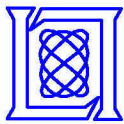
- SINR loss is a useful diagnostic, but it does not always translate directly into what we care about: detection performance
  - Case in point: SINR loss for stationary targets is infinite, but SAR detects stationary targets quite well!
- WL-GMTI straddles the regime between GMTI-type detection and SAR-type detection
- Need to consider detection theory to understand true capabilities of WL-GMTI



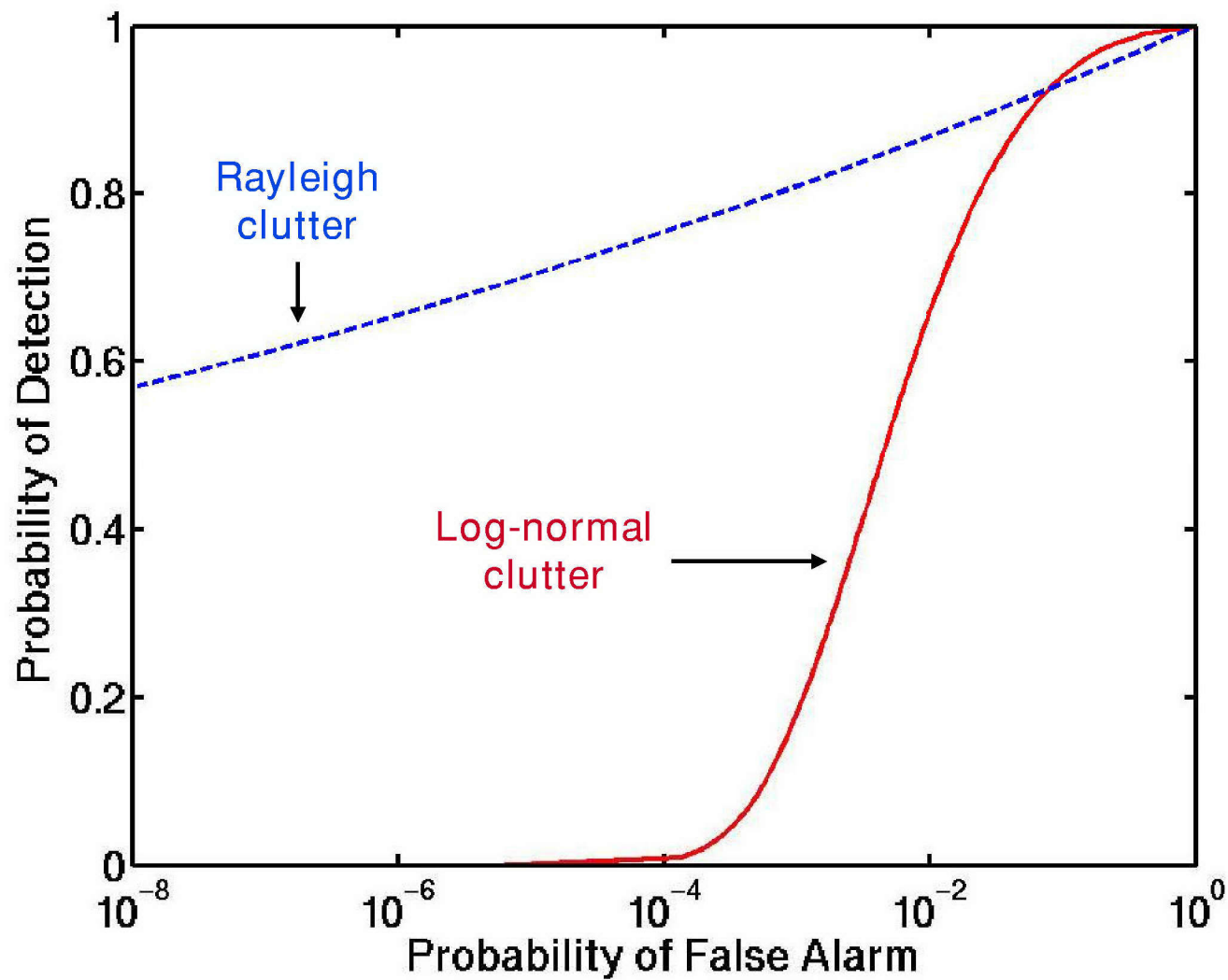
# Illustrative Toy Model

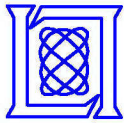
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- Radar: single-phase center (pure SAR)
- Moving point target with Rayleigh fading
- 15 dB mean target-to-clutter (for focused stationary target)
- Two clutter models:
  - Rayleigh (unrealistic, weak tails)
  - Log-normal (more realistic, heavier tails)

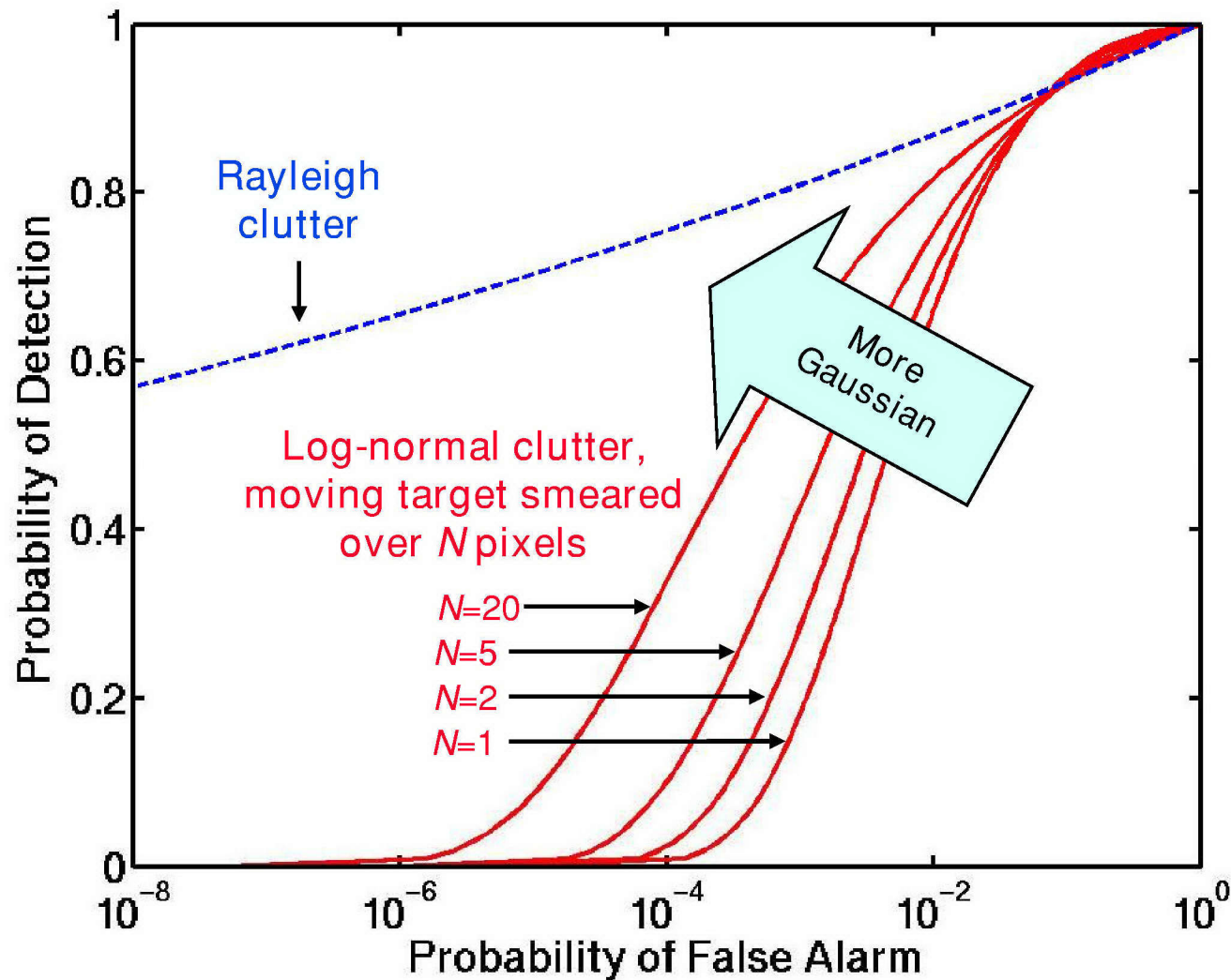


# Stationary Target Detection

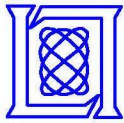




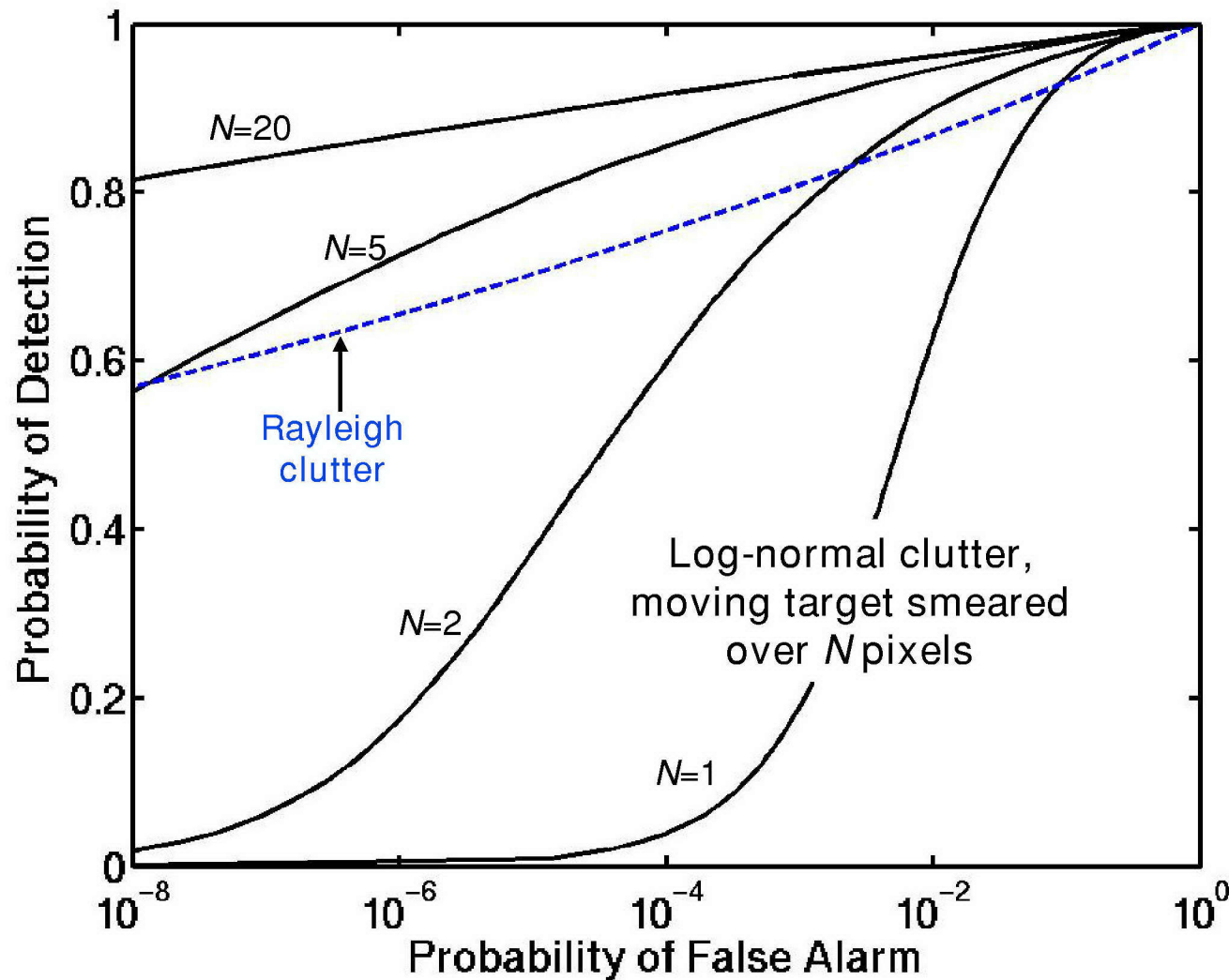
# Moving Target: Matched Filter Detector



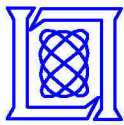




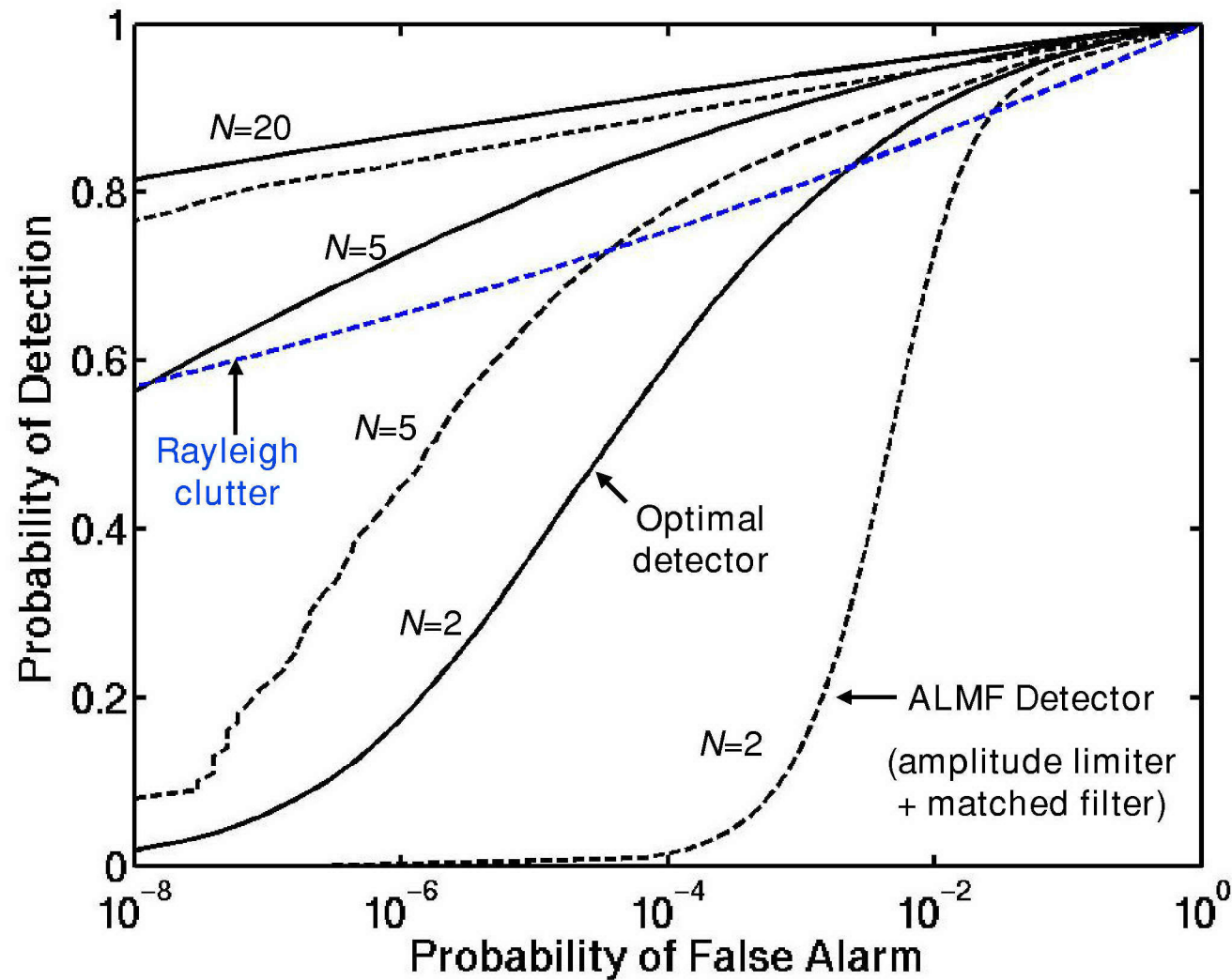
# Moving Target: Bayes-Optimal Detector

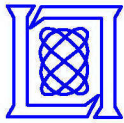






# Moving Target: ALMF Sub-Optimal Detector





# Summary

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- Wideband, long-CPI methods offers the promise of detecting slow, low-RCS targets not detectable with traditional GMTI methods
- This talk has explored some basic building blocks for analysis
  - Wideband, long-CPI data model and steering vectors
  - SINR loss analysis
- It appears that the detection capability of WL-GMTI straddles the SAR and GMTI domains: SINR loss alone is not a reliable metric of performance
  - Smearing of target over many pixels can enhance detection in strong-tailed clutter
  - Sub-optimal detector can approach optimal bound
- Many other aspects of the problem are ripe to be explored